

# ARTIFICIAL INTELLIGENCE

# Revision

simple reflex agents

reflex agents with state

goal-based agents

utility-based agents

All of these 4 types can be turned into learning agents

# Problem Solving

Problem-Solving Agents -> Atomic Representation

Planning Agents -> Factored or Structured Representation

**Assumption:** The solution of a problem is a fixed sequence of actions and this sequence DOES NOT depend upon future percepts

# Uninformed Search vs. Informed Search

**uninformed** search algorithms—algorithms that are given no information about the problem other than its definition.

**Informed** search algorithms, on the other hand, can do quite well given some guidance on where to look for solutions

# Problem Solving Agent

- 1) Goal Formulation
- 2) Problem Formulation
- 3) Searching Solution
- 4) Executing Solution

# 1. Goal Formulation

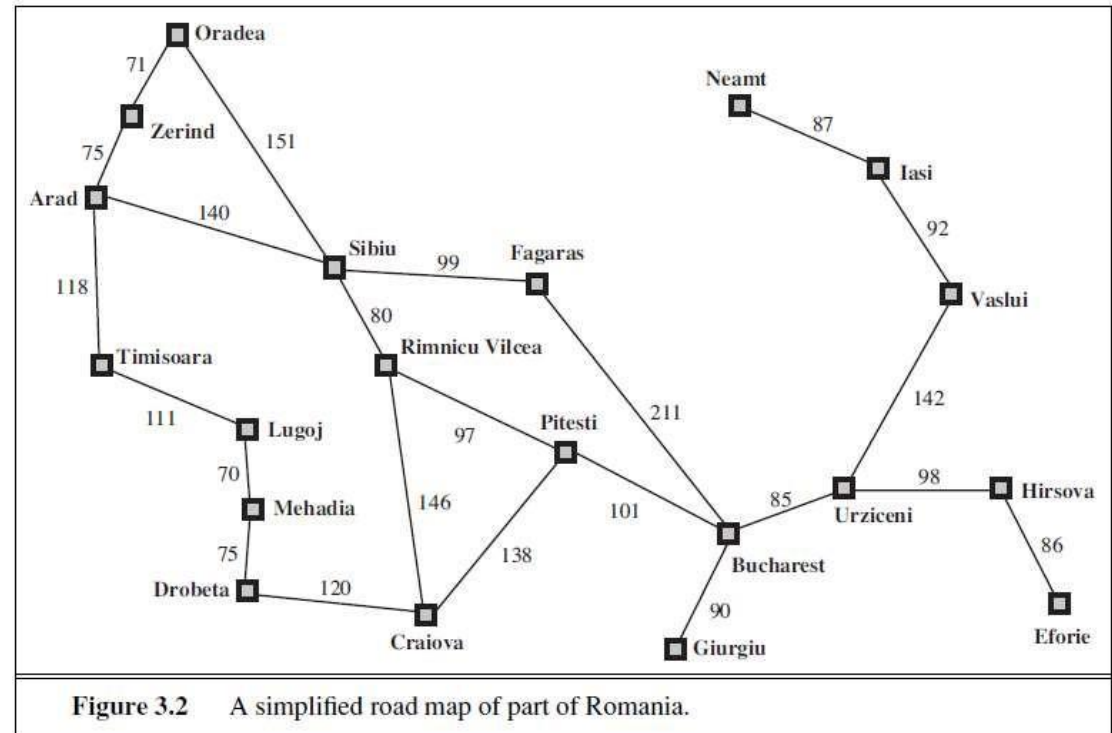
Goal formulation, based on the **current situation/state** and the agent's **performance measure**, is the first step in problem solving.

Performance Measure: Visit as many cities as possible, spend as low as possible on fuel.

Current State: In Arad



Possible Goal: In Bucharest



# Problem Formulation

**Problem formulation** is the process of deciding what actions and states to consider, given a goal.

The agent will consider actions at the level of driving from one major town to another. Each state therefore corresponds to being in a particular town.

Go Left, Go Right, Go forward, Go reverse. If these 4 actions are considered agent will never go out of the parking lot let alone reaching Bucharest.

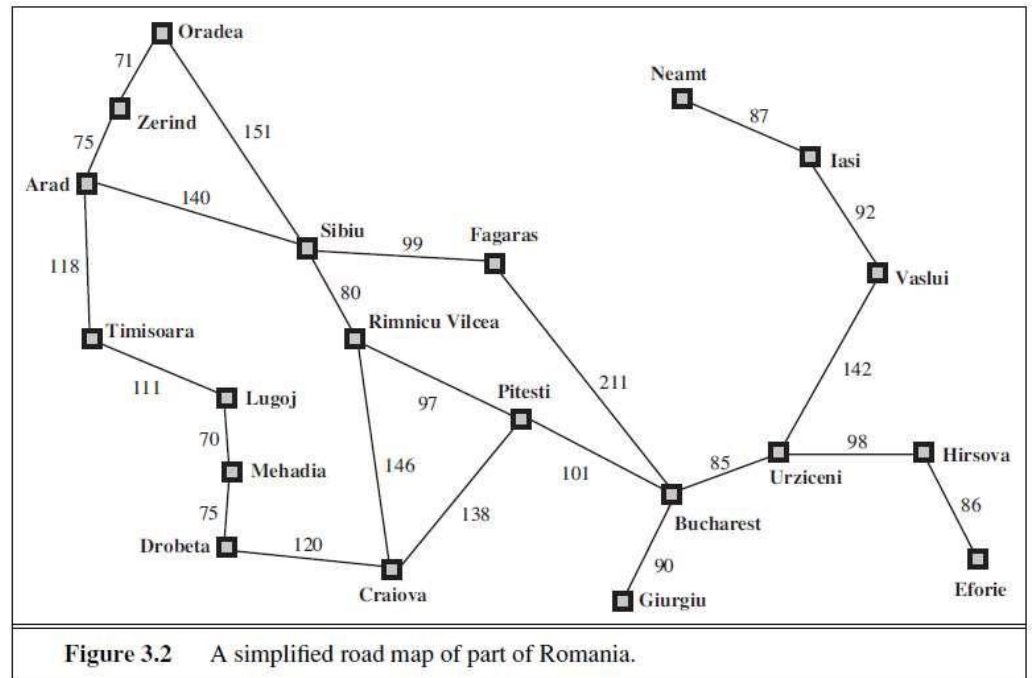
Possible Action: Go Cityname

# State Space Graph

On the right is the state space graph of our problem

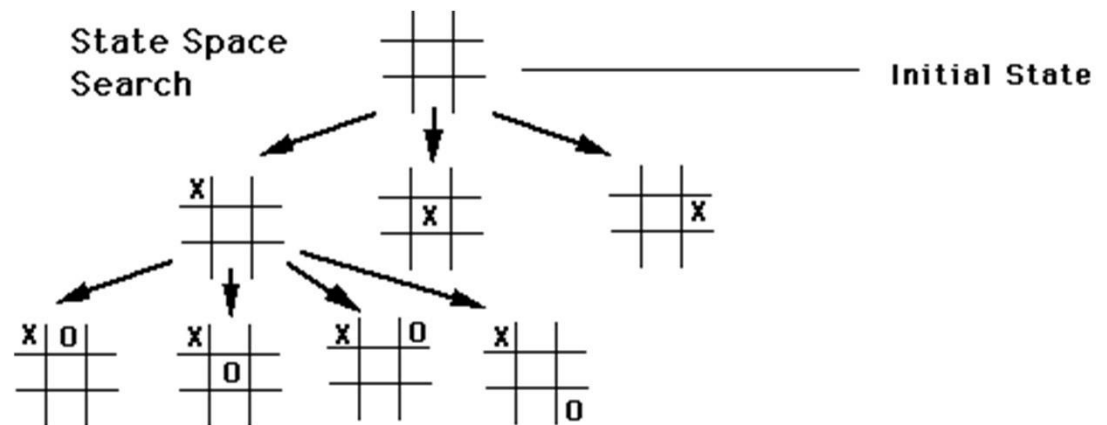
**State Space** - the set of all states reachable from the initial state by any sequence of actions.

A **path** in the state space is a sequence of states connected by a sequence of actions.





# State Space Graph



How would a state space of chess look like?

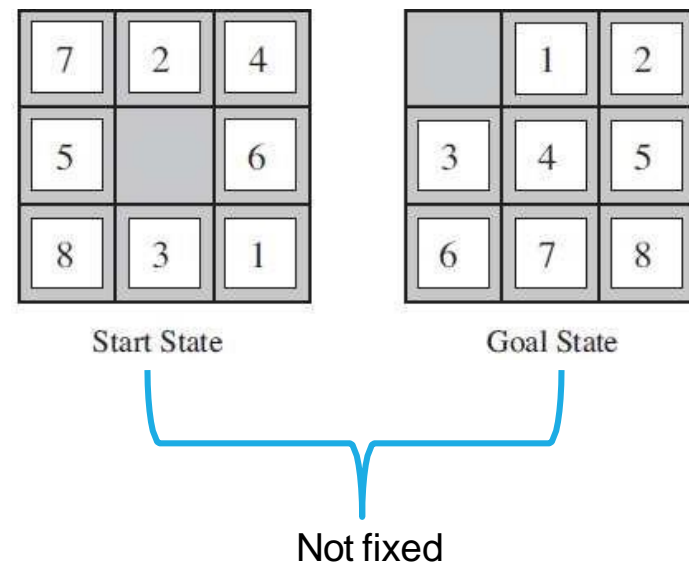
# Toy Problem: 8 Puzzle

Any state can be used as starting and Goal states.

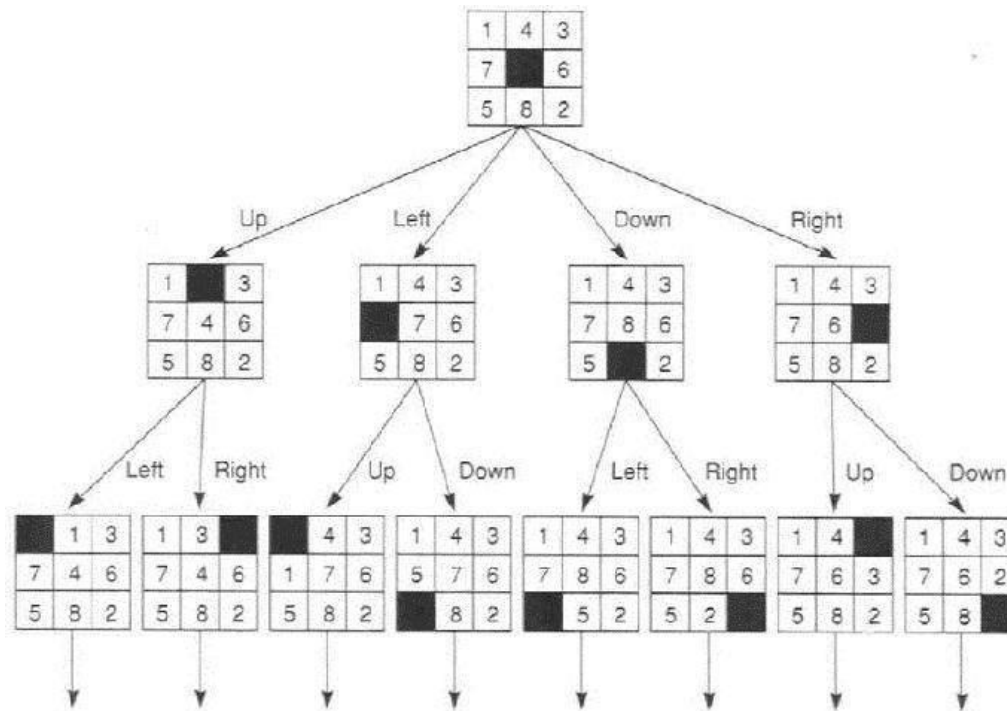
Actions: *Left*, *Right*, *Up*, or *Down*.

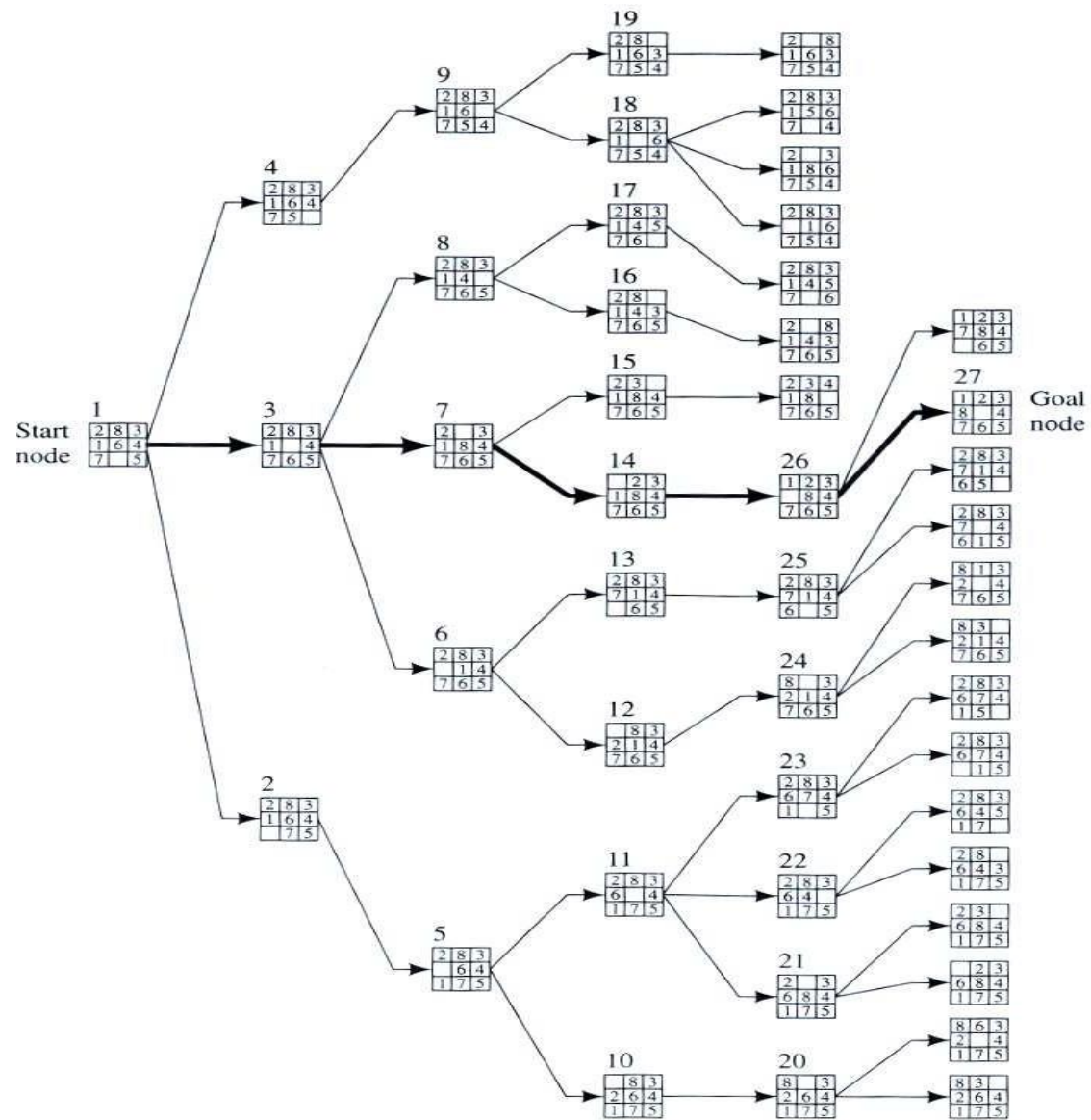
**Transition model:** Given a state and action, this returns the resulting state; for example, if we apply *Left* to the start state in adjacent figure, the resulting state has the 5 and the blank switched.

**Path Cost:** Each step costs 1, so the path cost is the number of steps in the path.



# Toy Problem: 8 Puzzle

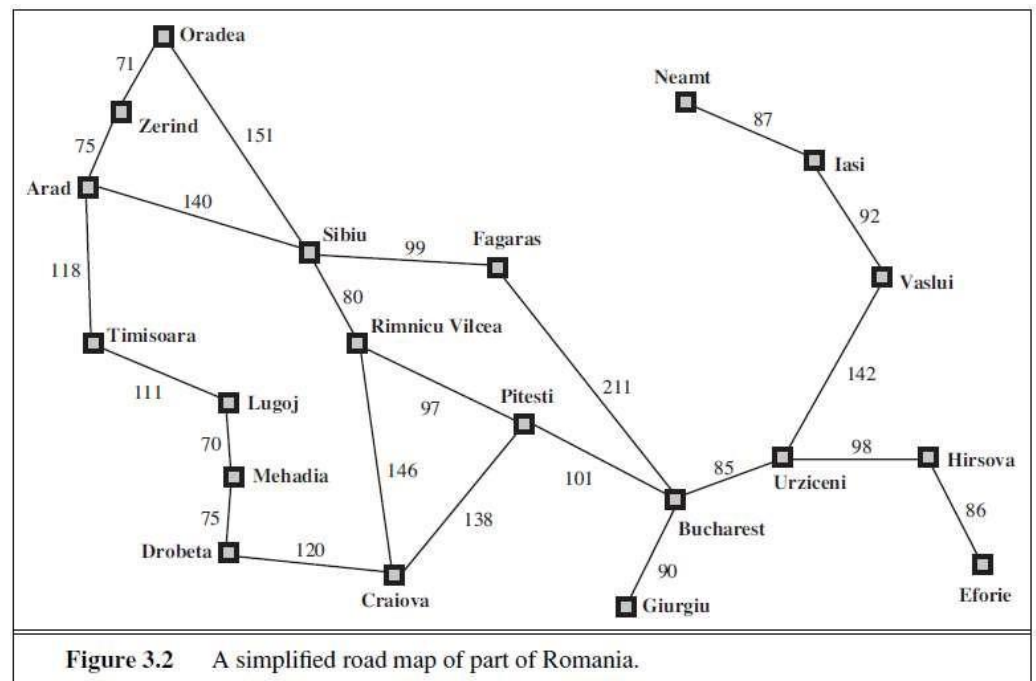




# Problem Definition

A problem consists of five components,

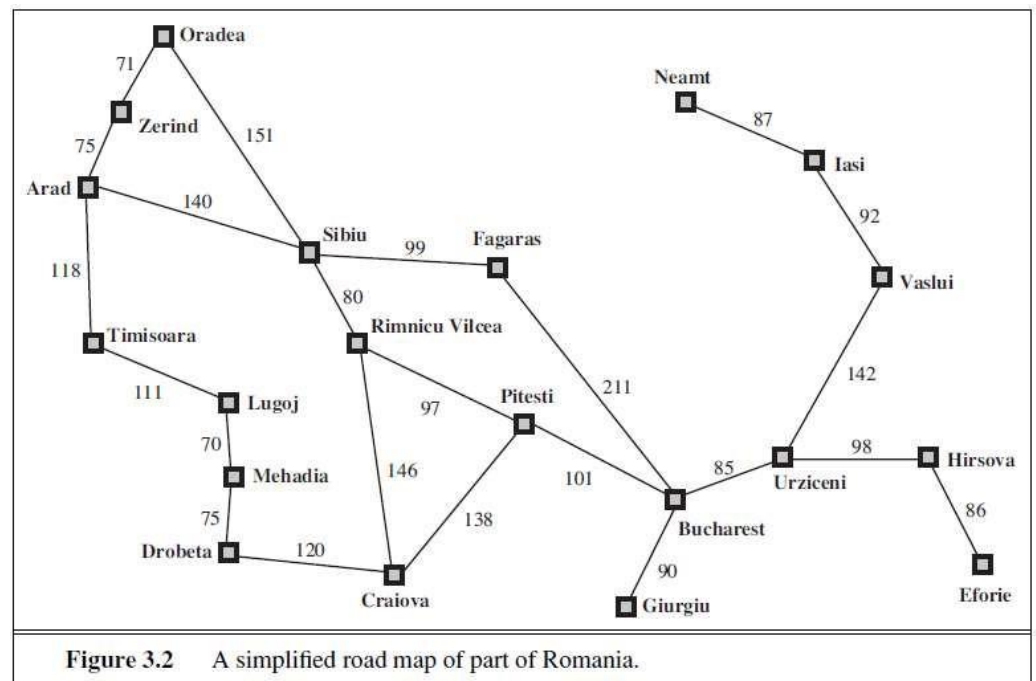
- Initial State
- Possible Actions
- Transition Model
- Goal Test
- Path Cost



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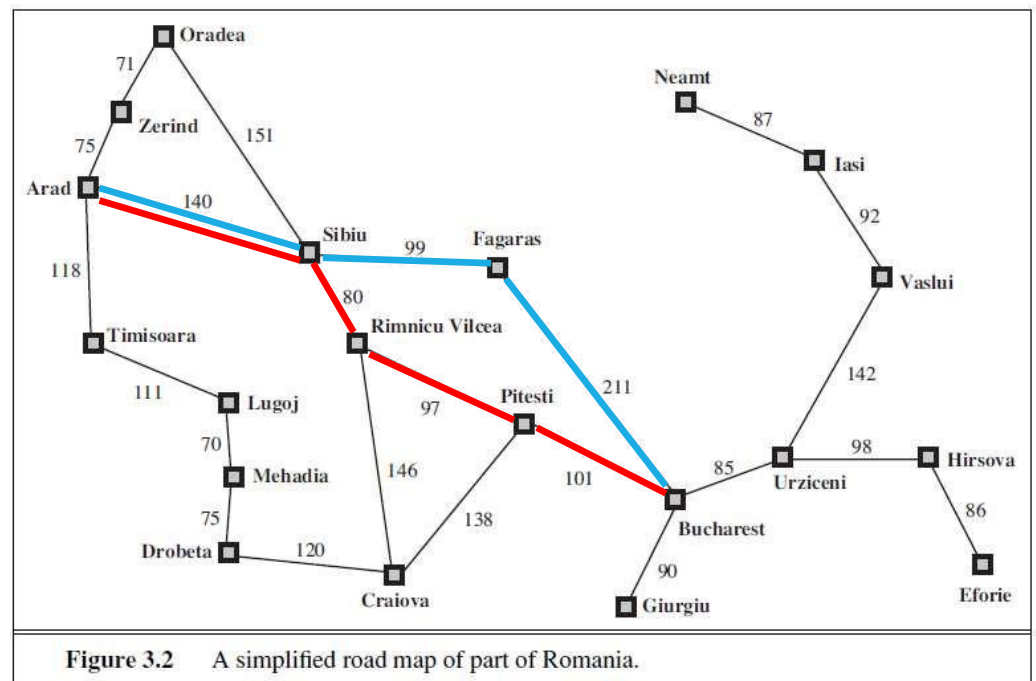
- Initial State : in (Arad)
- Possible Actions (given a state): e.g., from Arad possible actions are {Go(Sibiu), Go(Timisoara), Go(Zerind)}.
- Transition Model: specifies the relationship between a state, a possible action and the resulting successor state e.g.,  
 $\text{RESULT}(\text{In}(\text{Arad}), \text{Go}(\text{Zerind})) = \text{In}(\text{Zerind})$



# Problem Definition

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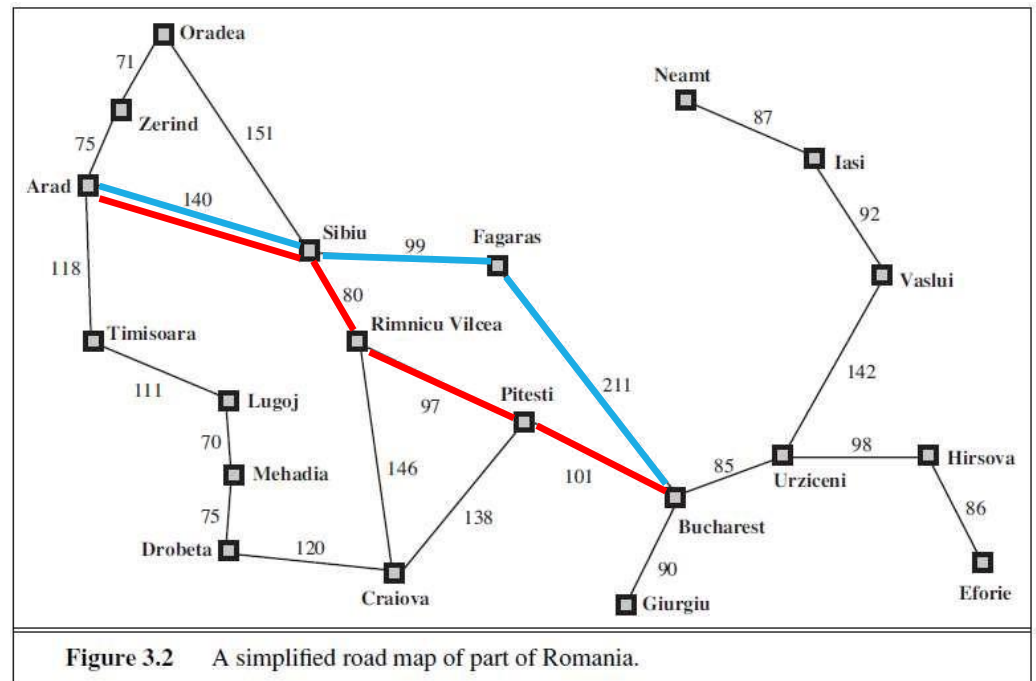
- Initial State
- Possible Actions
- Transition Model
- Goal Test (whether a given state is a goal state) e.g.  $\{ \text{In}(\text{Bucharest}) \}$ .
- Path Cost (based upon agent's performance measure). Two paths shown on the right. **Step cost** (the cost of a single action within a path)



# Solution Definition

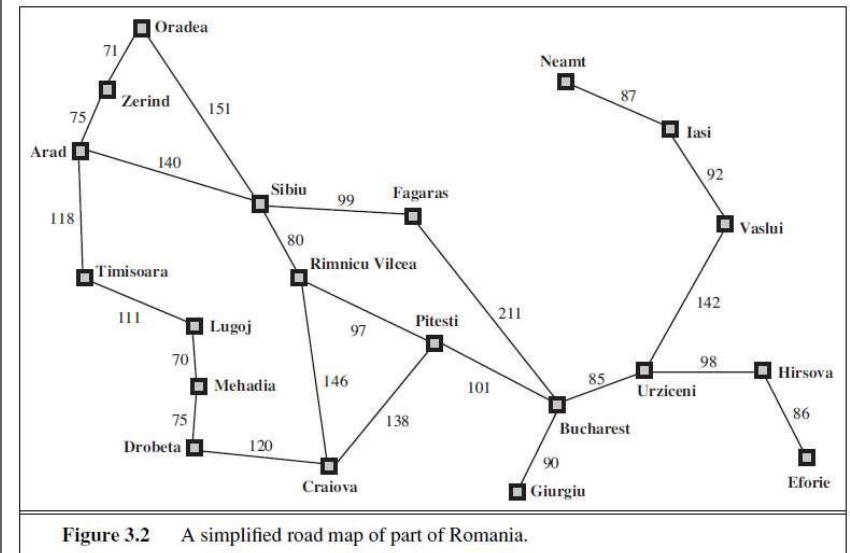
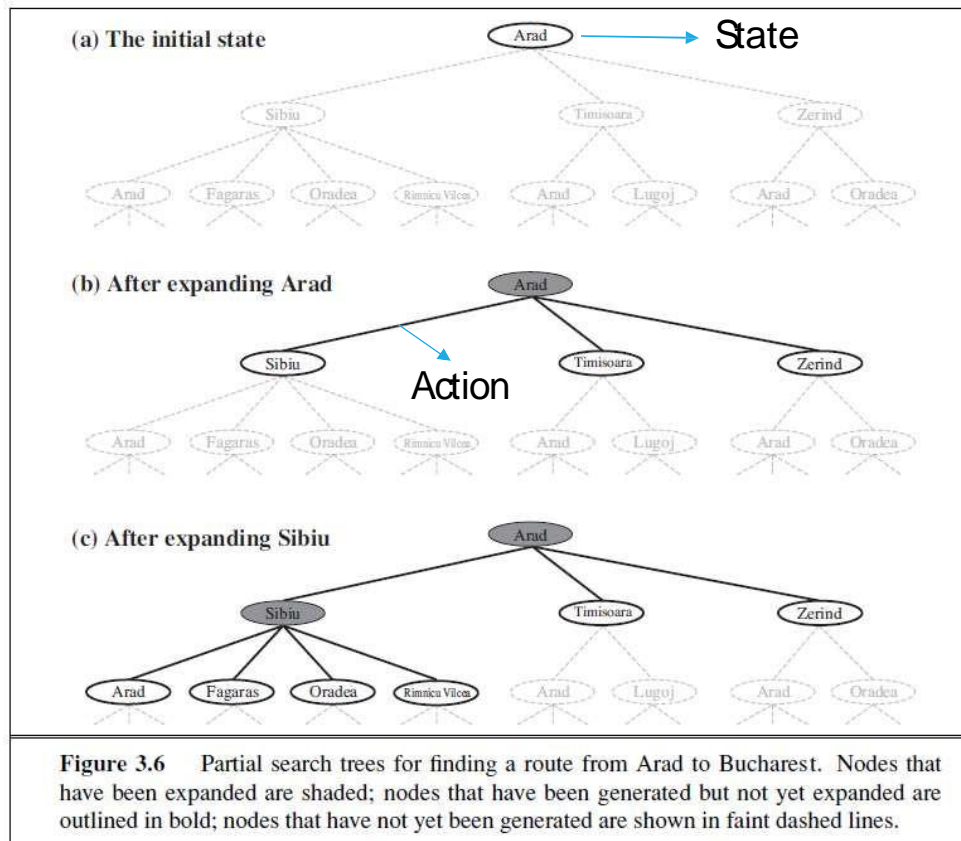
A **solution** to a problem is an action sequence that leads from the initial state to a goal state.

Solution quality is measured by the path cost function, and an **optimal solution** has the lowest path cost among all solutions.



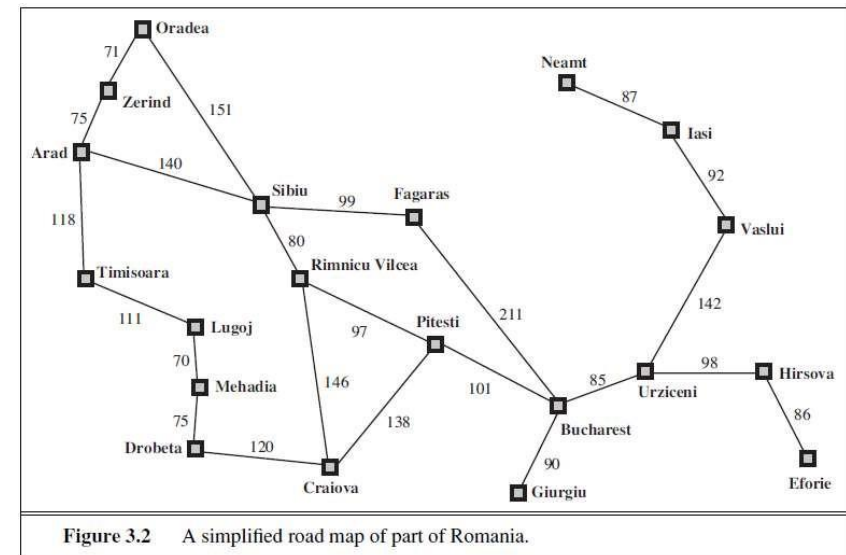
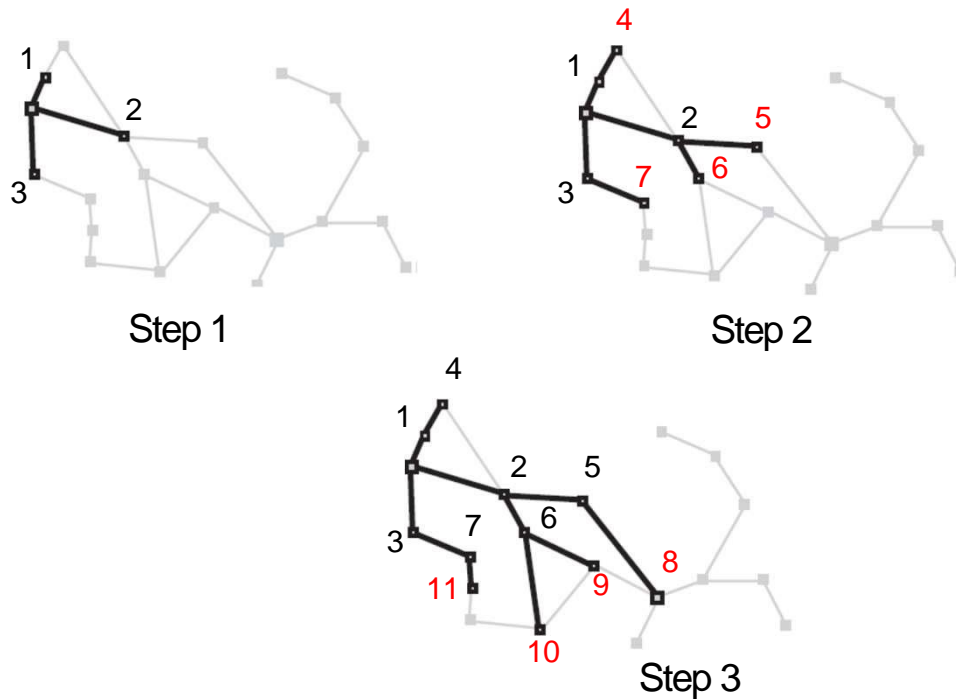


# Tree Search



# Graph Search

Use an **explore set** / **closed list** / **frontier** to remember the states already visited



# Infrastructure for Search Algorithms

- $n$ .STATE: the state in the state space to which the node corresponds;
- $n$ .PARENT: the node in the search tree that generated this node;
- $n$ .ACTION: the action that was applied to the parent to generate the node;
- $n$ .PATH-COST: the cost, traditionally denoted by  $g(n)$ , of the path from the initial state to the node, as indicated by the parent pointers.

Frontier / expanded nodes / unexpanded nodes are stored in the form of a queue (can be FIFO, LIFO (stack) or priority queue)

# Measuring Problem-solving Performance

**Completeness:** Is the algorithm guaranteed to find a solution when there is one?

**Optimality:** Does the strategy find the optimal solution.

**Time complexity:** How long does it take to find a solution?

**Space complexity:** How much memory is needed to perform the search?

# Measuring Problem-solving Performance

Complexity in a tree or a graph is measured by:

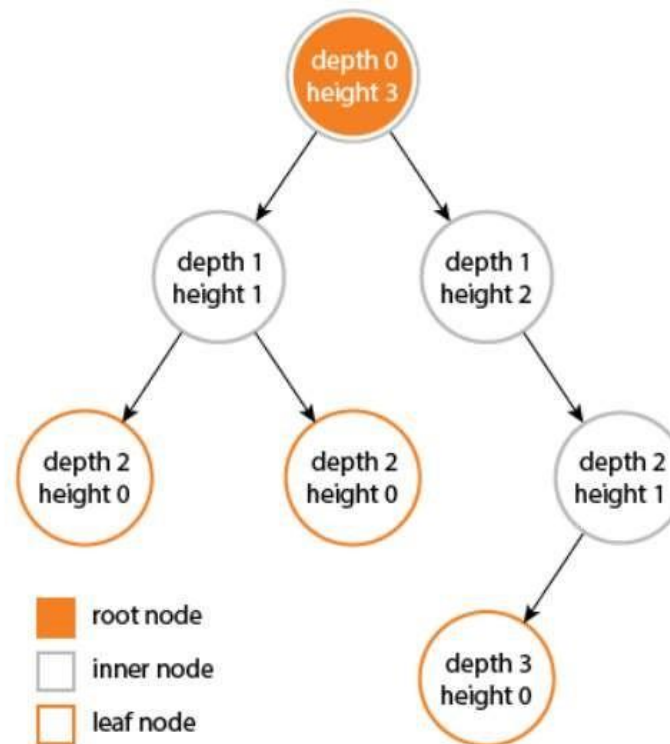
- 1)  $b$ , the **branching factor** or maximum number of children of any node
- 2)  $d$ , the **depth** of the goal node (i.e., the number of steps along the path from the root)
- 3)  $m$ , the maximum possible depth of entire tree.

Time is often measured in terms of the number of nodes generated during the search.

Space in terms of the maximum number of nodes stored in memory.

# Measuring Problem-solving Performance

A binary tree has a branching factor  $c$



# Complexity

Branching Factor,  $b=3$  (Max num of children for any node is 3)

Depth,  $d = 3$  (Bucharest is 3 steps away from Arad)

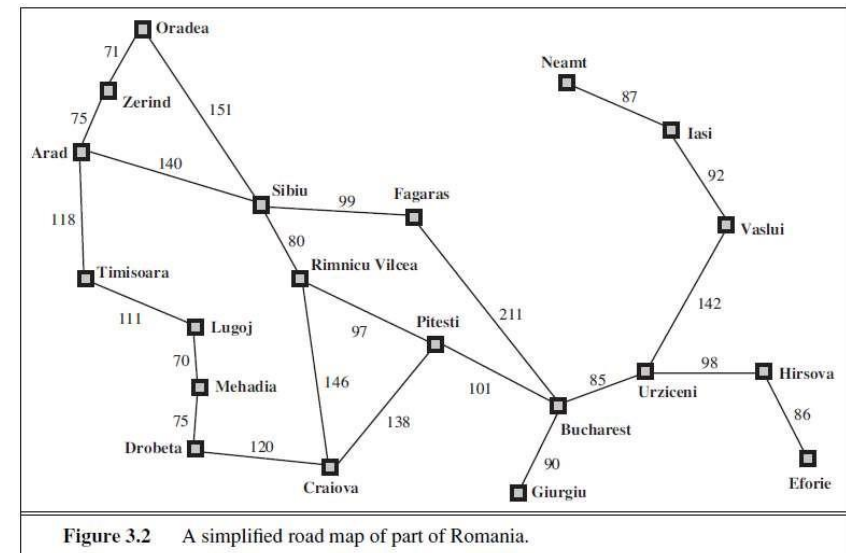
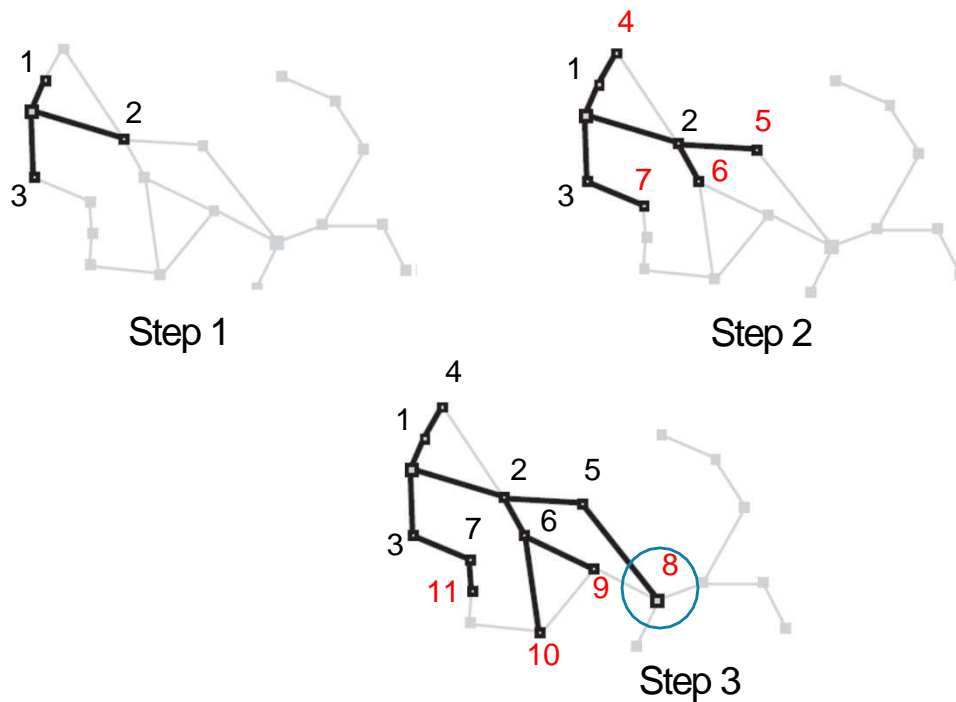
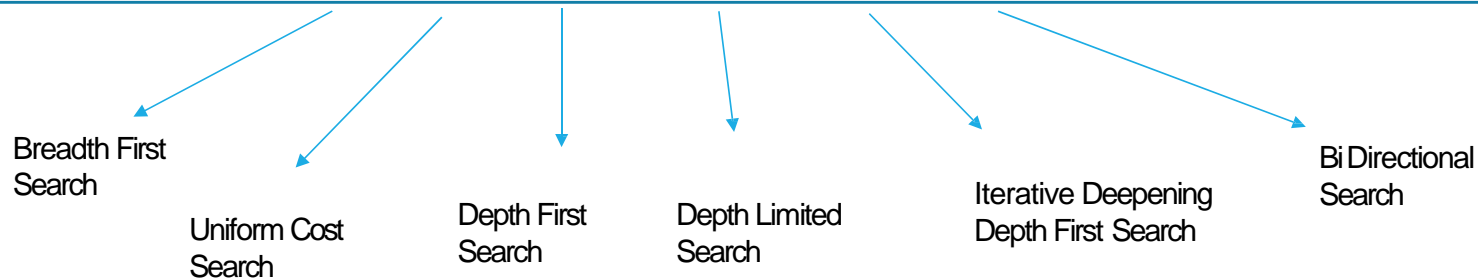


Figure 3.2 A simplified road map of part of Romania.

# Informed vs Uninformed Search

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# Breadth First Search

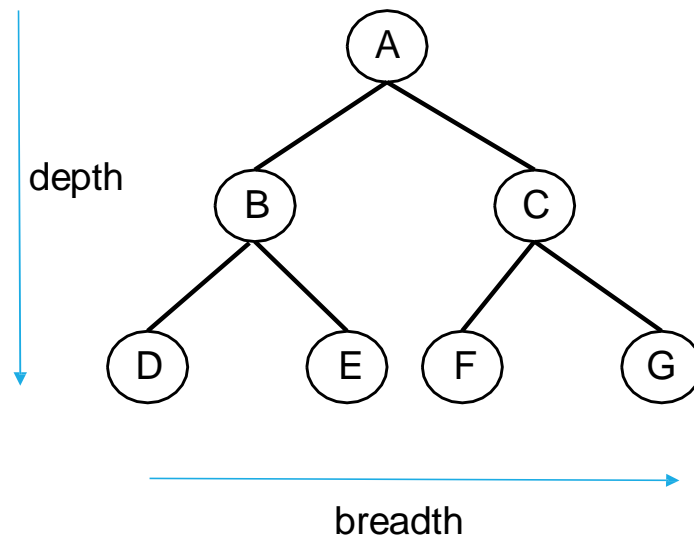
Expand Shallowest Node First

*Frontier* (or fringe): nodes in queue to be explored

Reached: Nodes that are already explored

*Frontier* is a first-in-first-out (FIFO) queue, i.e., new successors go at end of the queue.

*Goal-Test* when inserted



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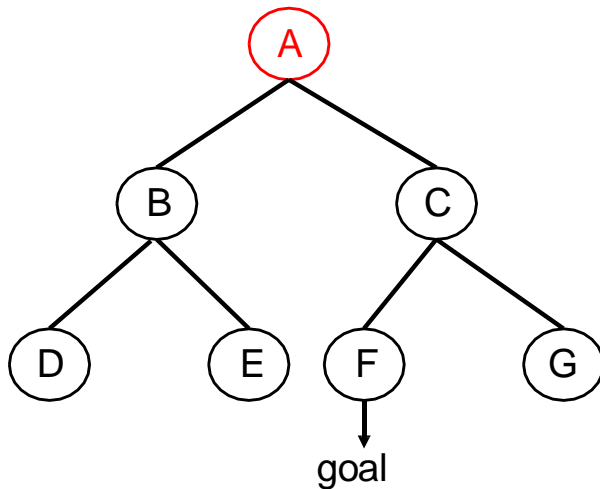
```
function BREADTH-FIRST-SEARCH(problem) returns a solution node or failure  
  node ← NODE(problem.INITIAL)  
  if problem.IS-GOAL(node.STATE) then return node  
  frontier ← a FIFO queue, with node as an element  
  reached ← {problem.INITIAL}  
  while not IS-EMPTY(frontier) do  
    node ← POP(frontier)  
    for each child in EXPAND(problem, node) do  
      s ← child.STATE  
      if problem.IS-GOAL(s) then return child  
      if s is not in reached then  
        add s to reached  
        add child to frontier  
  return failure
```

# Breadth First Search

Is A a goal state?

Frontier = [A]

Reached = [ ]



**function** BREADTH-FIRST-SEARCH(*problem*) **returns** a solution node or *failure*

*node* ← NODE(*problem*.INITIAL)

**if** *problem*.IS-GOAL(*node*.STATE) **then return** *node*

*frontier* ← a FIFO queue, with *node* as an element

*reached* ← {*problem*.INITIAL}

**while not** IS-EMPTY(*frontier*) **do**

*node* ← POP(*frontier*)

**for each** *child* **in** EXPAND(*problem*, *node*) **do**

*s* ← *child*.STATE

**if** *problem*.IS-GOAL(*s*) **then return** *child*

**if** *s* is not in *reached* **then**

add *s* to *reached*

add *child* to *frontier*

**return** *failure*

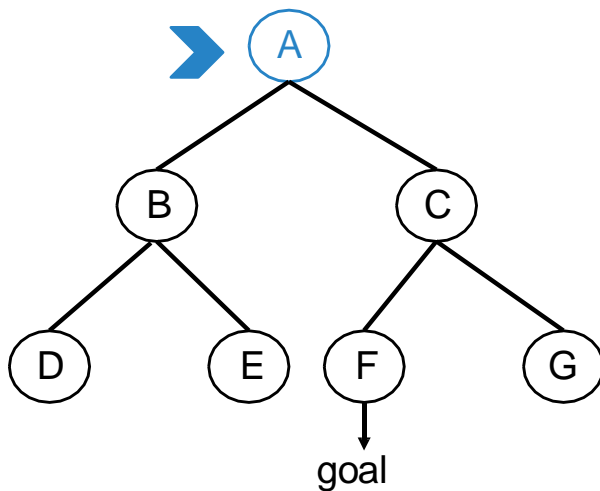
# Breadth First Search

Frontier = [A]

Reached = [ ]

Frontier = [ ]

Reached = [ A ]



**function** BREADTH-FIRST-SEARCH(*problem*) **returns** a solution node or *failure*

*node* ← NODE(*problem*.INITIAL)

**if** *problem*.IS-GOAL(*node*.STATE) **then return** *node*

*frontier* ← a FIFO queue, with *node* as an element

*reached* ← {*problem*.INITIAL}

**while not** IS-EMPTY(*frontier*) **do**

*node* ← POP(*frontier*)

**for each** *child* **in** EXPAND(*problem*, *node*) **do**

*s* ← *child*.STATE

**if** *problem*.IS-GOAL(*s*) **then return** *child*

**if** *s* is not in *reached* **then**

add *s* to *reached*

add *child* to *frontier*

**return** *failure*

➡ Node

➡ Child

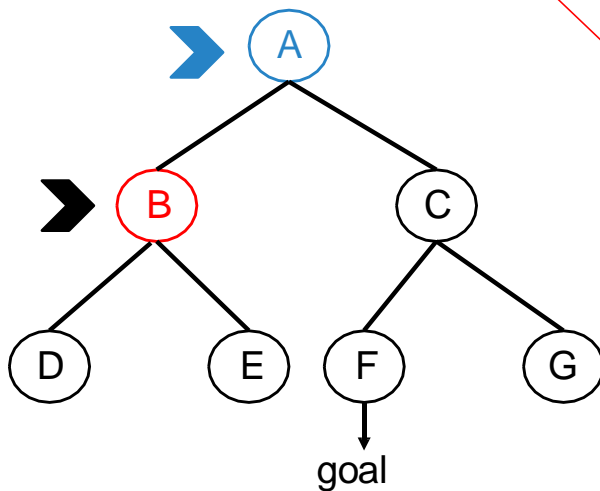
# Breadth First Search

Frontier = [ ]

Reached = [ A ]

Frontier = [ B ]

Reached = [ A ]



```
function BREADTH-FIRST-SEARCH(problem) returns a solution node or failure
  node ← NODE(problem.INITIAL)
  if problem.IS-GOAL(node.STATE) then return node
  frontier ← a FIFO queue, with node as an element
  reached ← {problem.INITIAL}
  while not IS-EMPTY(frontier) do
    node ← POP(frontier)
    for each child in EXPAND(problem, node) do
      s ← child.STATE
      if problem.IS-GOAL(s) then return child
      if s is not in reached then
        add s to reached
        add child to frontier
  return failure
```

Is B a goal state?

➡ Node

➡ Child

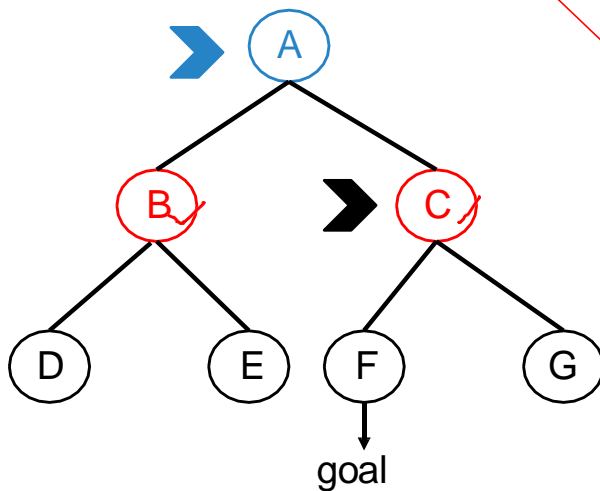
# Breadth First Search

Frontier = [ B ]

Reached = [ A ]

Frontier = [ B, C ]

Reached = [ A ]



```
function BREADTH-FIRST-SEARCH(problem) returns a solution node or failure
  node ← NODE(problem.INITIAL)
  if problem.IS-GOAL(node.STATE) then return node
  frontier ← a FIFO queue, with node as an element
  reached ← {problem.INITIAL}
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    node ← POP(frontier)
    for each child in EXPAND(problem, node) do
      s ← child.STATE
      if problem.IS-GOAL(s) then return child
      if s is not in reached then
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        add child to frontier
  return failure
```

Is C a goal state?

➡ Node

➡ Child

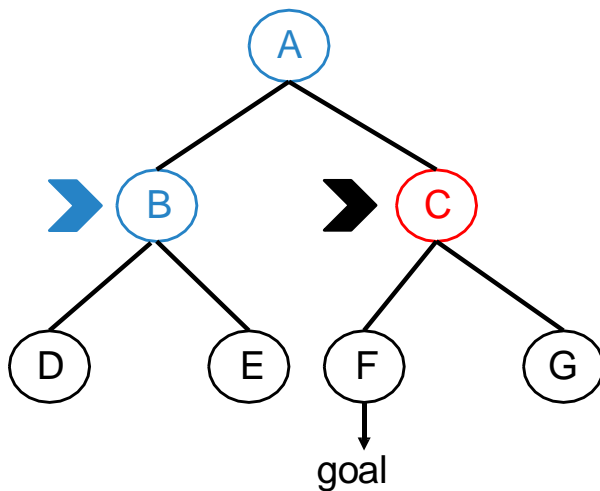
# Breadth First Search

Frontier = [ B, C ]

Reached = [ A ]

Frontier = [ C ]

Reached = [ A, B ]



```
function BREADTH-FIRST-SEARCH(problem) returns a solution node or failure
  node ← NODE(problem.INITIAL)
  if problem.IS-GOAL(node.STATE) then return node
  frontier ← a FIFO queue, with node as an element
  reached ← {problem.INITIAL}
  while not IS-EMPTY(frontier) do
    node ← POP(frontier)
    for each child in EXPAND(problem, node) do
      s ← child.STATE
      if problem.IS-GOAL(s) then return child
      if s is not in reached then
        add s to reached
        add child to frontier
  return failure
```

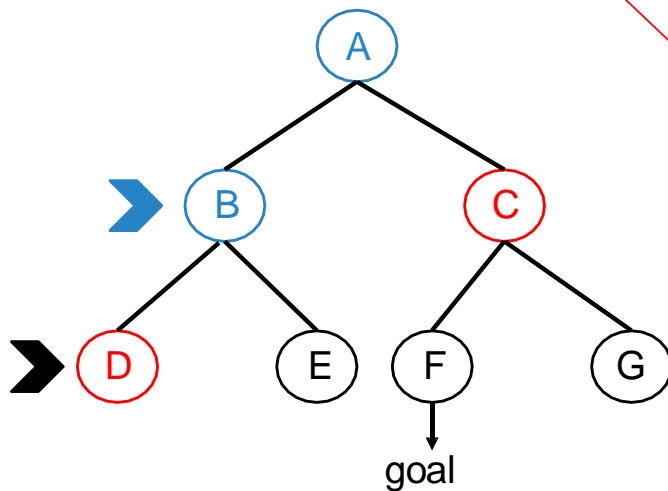
➡ Node

➡ Child

# Breadth First Search

Frontier = [ C ]  
Reached = [ A, B ]

Frontier = [ C, D ]  
Reached = [ A, B ]



```
function BREADTH-FIRST-SEARCH(problem) returns a solution node or failure
  node ← NODE(problem.INITIAL)
  if problem.IS-GOAL(node.STATE) then return node
  frontier ← a FIFO queue, with node as an element
  reached ← {problem.INITIAL}
  while not IS-EMPTY(frontier) do
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    for each child in EXPAND(problem, node) do
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      if problem.IS-GOAL(s) then return child
      if s is not in reached then
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        add child to frontier
  return failure
```

Is D a goal state?

➡ Node

➡ Child



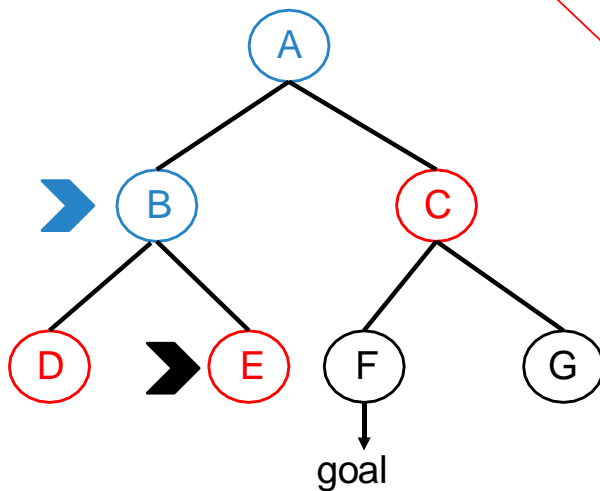
# Breadth First Search

Frontier = [ C, D ]

Reached = [ A, B ]

Frontier = [ C, D, E ]

Reached = [ A, B ]



```
function BREADTH-FIRST-SEARCH(problem) returns a solution node or failure
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  reached ← {problem.INITIAL}
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    node ← POP(frontier)
    for each child in EXPAND(problem, node) do
      s ← child.STATE
      if problem.IS-GOAL(s) then return child
      if s is not in reached then
        add s to reached
        add child to frontier
  return failure
```

Is E a goal state?

➡ Node

➡ Child

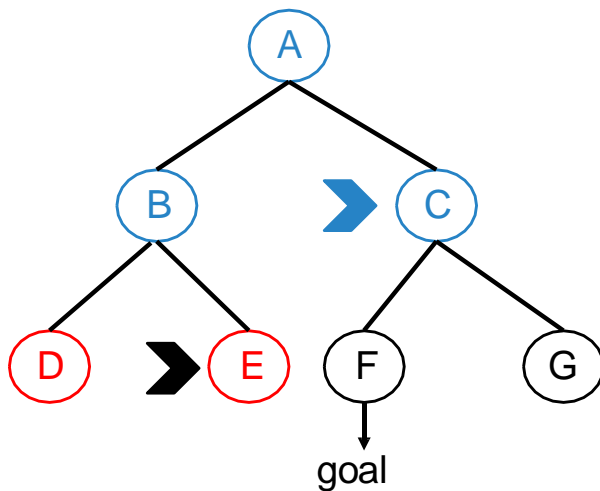
# Breadth First Search

Frontier = [ C, D, E ]

Reached = [ A, B ]

Frontier = [ D, E ]

Reached = [ A, B, C ]



```
function BREADTH-FIRST-SEARCH(problem) returns a solution node or failure
  node ← NODE(problem.INITIAL)
  if problem.IS-GOAL(node.STATE) then return node
  frontier ← a FIFO queue, with node as an element
  reached ← {problem.INITIAL}
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      if s is not in reached then
        add s to reached
        add child to frontier
  return failure
```

➤ Node

➤ Child

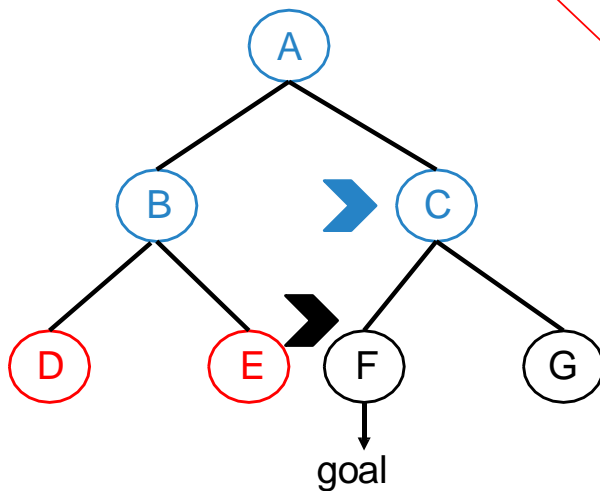
# Breadth First Search

Frontier = [ D, E ]

Reached = [ A, B, C ]

Frontier = [ D, E ]

Reached = [ A, B, C ]



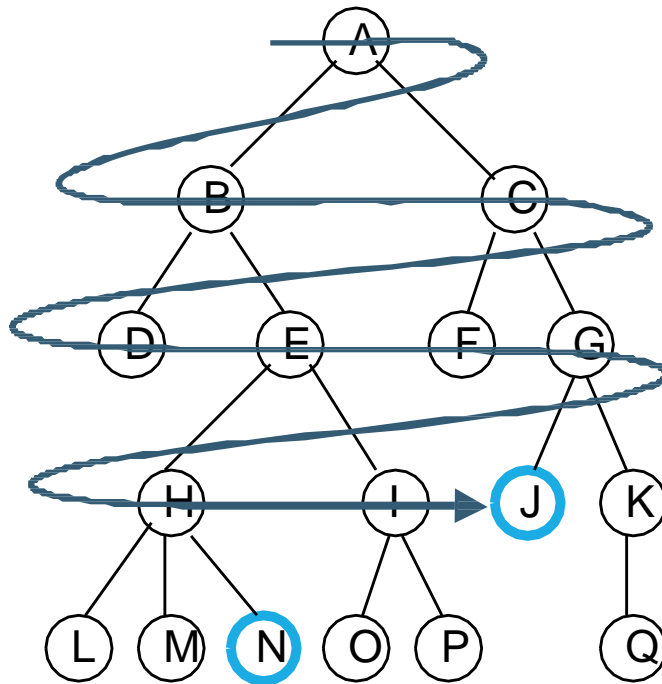
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    for each child in EXPAND(problem, node) do
      s ← child.STATE
      if problem.IS-GOAL(s) then return child
      if s is not in reached then
        add s to reached
        add child to frontier
  return failure
```

Is F a goal state?

➤ Node

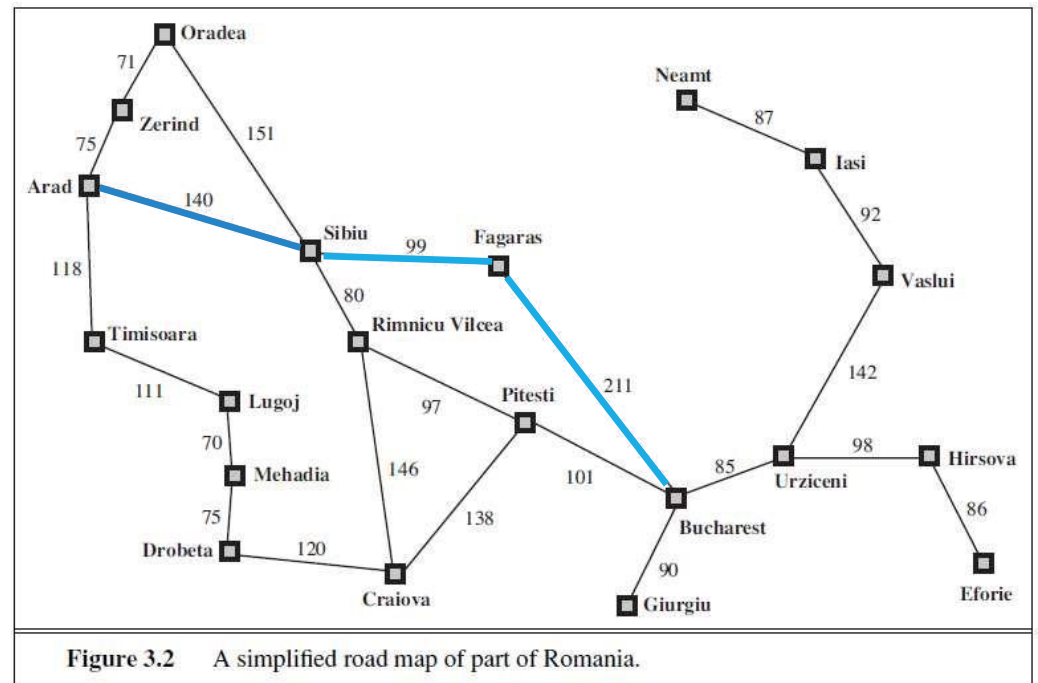
➤ Child

# BFS: Summary



# BFS: Completeness

if the shallowest goal node is at some **finite depth d**, breadth-first search will eventually find it after generating all shallower nodes (provided the branching factor **b is finite**) .



BFSgoal path

# BFS: Optimal?

Not necessarily optimal

Only optimal if every action has same cost.

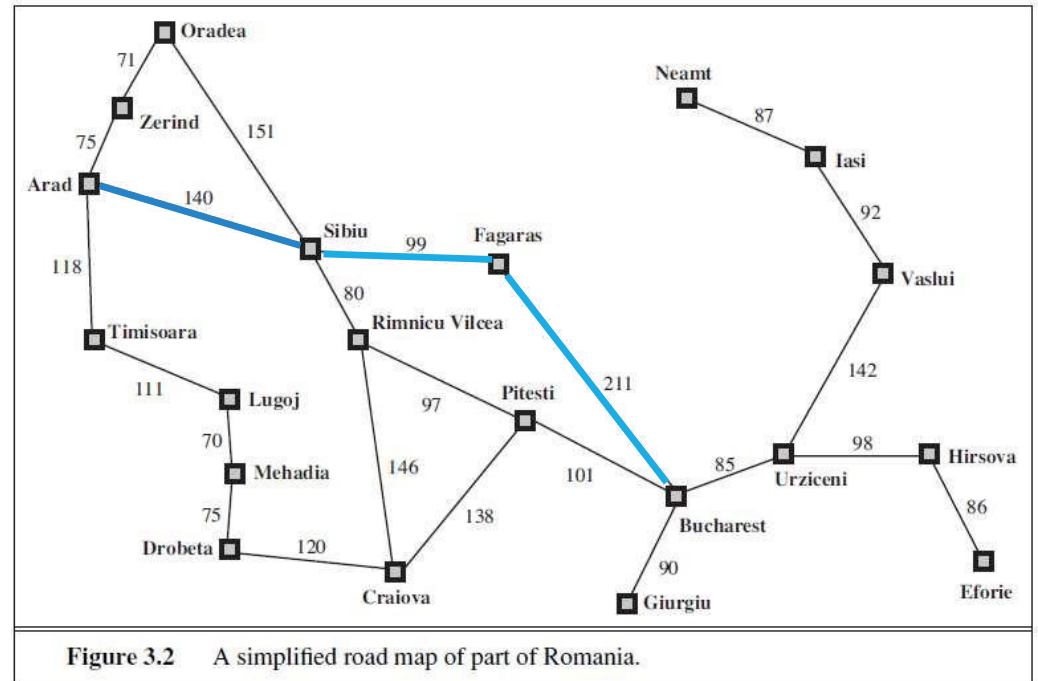


Figure 3.2 A simplified road map of part of Romania.

BFSgoal path

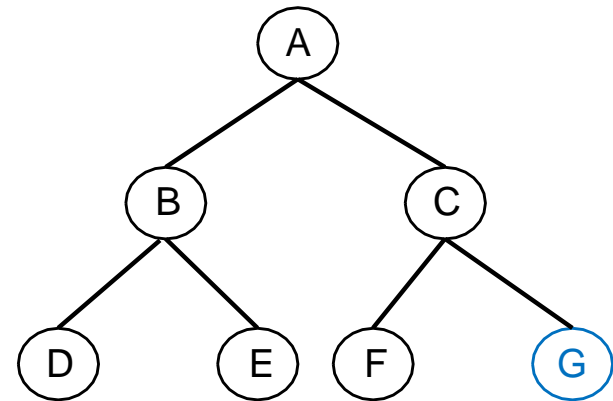
# BFS: Time Complexity

Worst complexity is when G is a goal state.

In this case, total number of nodes generated is

$b + b^2 + b^3 + \dots + b^d = O(b^d)$  (The  $d^{\text{th}}$  layer contains nodes much larger than all the nodes in previous layers combined!)

So the time complexity is  $O(b^d)$



A binary tree,  $b=2$ ,  $d=2$

# BFS: Space Complexity

There will be  $O(b^{d-1})$  nodes in the explored set and  $O(b^d)$  nodes in the frontier.

So the space complexity is  $O(b^d)$ , i.e., it is dominated by the size of the frontier.

Exponential time complexity can be accepted but exponential space complexity is BAD !

